

# KOLMOGOROV COMPLEXITY AND THE ASYMPTOTIC BOUND FOR ERROR-CORRECTING CODES

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*To the memory of Friedrich Hirzebruch, a great mathematician and friend*

## Abstract

The set of all error-correcting block codes over a fixed alphabet with  $q$  letters determines a recursively enumerable set of rational points in the unit square with coordinates  $(R, \delta) := (\text{relative transmission rate}, \text{relative minimal distance})$ . Limit points of this set form a closed subset, defined by  $R \leq \alpha_q(\delta)$ , where  $\alpha_q(\delta)$  is a continuous decreasing function called the *asymptotic bound*. Its existence was proved by the first-named author in 1981 ([10]), but no approaches to the computation of this function are known, and in [14] it was even suggested that this function might be uncomputable in the sense of constructive analysis.

In this note we show that the asymptotic bound becomes computable with the assistance of an oracle producing codes in the order of their growing Kolmogorov complexity. Moreover, a natural partition function involving complexity allows us to interpret the asymptotic bound as a curve dividing two different thermodynamic phases of codes.

## 1. Introduction

In this article, we address again two related basic problems about asymptotic bounds for codes, discussed recently in [14] and in [15]. The first one is the problem of *computability*, or, more suggestively, *plottability* of the bound. The second one is the problem of interpretation of this bound as a kind of phase-transition curve.

We start Section 2 below with precise definitions of the relevant notions, and the reader may wish to turn to it immediately. Here we restrict ourselves to some explanations on the intuitive level.

Consider all error-correcting block codes  $C$  in a fixed alphabet  $A$  with  $q$  letters. Each such code determines its *code point*  $(R(C), \delta(C))$  in the plane ( $R := \text{transmission rate}$ ,  $\delta := \text{minimal relative Hamming distance}$ ).

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The *asymptotic bound*  $R = \alpha_q(\delta)$  is a continuous curve in the plane  $(R, \delta)$  such that all limit points of the set of code points lie below or on this bound, whereas all isolated code points lie above it.

Since 1981, when the existence of the asymptotic bound (and its versions for various structured codes, such as linear ones) was discovered in [10], many estimates for it from above and from below were established, but no exact formula was found. This led one of us to conjecture in [14] that the function  $R = \alpha_q(\delta)$  might be *uncomputable* (and its graph unplotable) in the technical sense formalized in [5] and [3].

Here we treat codes assuming that *an oracle is given* that produces them in the order of *increasing Kolmogorov complexity*, and show that with the assistance of such an oracle  $R = \alpha_q(\delta)$  becomes “plottable” (Section 3), and that appropriate partition functions involving either the Kolmogorov complexity or Levin’s prefix complexity ([7]) change behavior across this asymptotic bound (Section 4). (At the end of [12] it was argued that civilization is such a universal oracle.)

Slightly more precisely, it is known (see [4], [1], [18], and Section 2 below) that if one chooses first a natural enumeration of codes and then generates error-correcting codes in the order of their “size” (actually, any computable order rather than complexity), then the density of the respective code points is concentrated below or near the bound (2.2) (for linear codes (2.3)) that lies in turn strictly below the asymptotic bound. Complexity is invoked here principally in order to identify typical random codes with codes whose complexity is comparable with their size: cf. [19], [8], [9].

By contrast, we show that code points of codes generated in the order of growing complexity, which puts a considerable amount of highly non-random codes at the foreground, tend to be well distributed below the asymptotic bound, with the bound itself appearing as a “silver lining” of the cloud of code points.

The last Section (5) is dedicated to a sketch of a “quantization” of the classical ensemble of codes.

## 2. Asymptotic bound as a non-statistical phenomenon

**2.1. Codes and code points.** Here we recall our main definitions and results from previous works.

We choose and fix an integer  $q \geq 2$  and a finite set, *the alphabet*  $A$ , or  $A_q$ , of cardinality  $q$ . An (unstructured) *code*  $C$  is defined as a nonempty subset  $C \subset A^n$  of words of length  $n \geq 1$ . Such  $C$  determines its *code point*

$$P_C = (R(C), \delta(C))$$

in the  $(R, \delta)$ -plane, where  $R(C)$  is called the *transmission rate* and  $\delta(C)$  is the *relative minimal distance of the code*. They are defined by the

formulas

$$\delta(C) := \frac{d(C)}{n(C)}, \quad d(C) := \min \{d(a, b) \mid a, b \in C, a \neq b\}, \quad n(C) := n,$$

$$(2.1) \quad R(C) = \frac{\lfloor k(C) \rfloor}{n(C)}, \quad k(C) := \log_q \text{card}(C),$$

where  $d(a, b)$  is the Hamming distance

$$d((a_i), (b_i)) := \text{card}\{i \in (1, \dots, n) \mid a_i \neq b_i\}.$$

In the degenerate case  $\text{card } C = 1$  we put  $d(C) = 0$ . We will call the numbers  $k = k(C)$ ,  $n = n(C)$ ,  $d = d(C)$ , *code parameters* and refer to  $C$  as an  $[n, k, d]_q$ -code. We denote by  $\text{Codes}_q$  the set of all such codes, and by  $cp : \text{Codes}_q \rightarrow [0, 1]^2 \cap \mathbb{Q}^2$  the map  $C \mapsto P_C$ . The *multiplicity* of a code point  $x$  is defined as cardinality of the fiber  $cp^{-1}(x)$ .

If  $q$  is a prime power, and  $A_q$  is endowed with a structure of a finite field  $\mathbb{F}_q$ , then a *linear code* is a linear subspace of  $A_q^n$ . The set of linear codes is denoted  $\text{Codes}_q^{\text{lin}}$ .

Our starting point here will be the following characterization of the set of all code points, first proved in its final form in [14]. Note that the  $R$ -axis is traditionally drawn as a vertical one.

**Theorem 2.1.** *There exists a continuous function  $\alpha_q(\delta)$ ,  $\delta \in [0, 1]$  such that*

- (i) *The set of code points of infinite multiplicity is exactly the set of rational points  $(R, \delta) \in [0, 1]^2$  satisfying  $R \leq \alpha_q(\delta)$ . The curve  $R = \alpha_q(\delta)$  is called the asymptotic bound.*
- (ii) *Code points  $x$  of finite multiplicity all lie above the asymptotic bound and are isolated: for each such point there is an open neighborhood containing  $x$  as the only code point.*
- (iii) *The same statements are true for linear codes, with a possibly different asymptotic bound  $R = \alpha_q^{\text{lin}}(\delta)$ .*

**2.2. Good codes.** One characteristic of a good code is this: *it maximizes simultaneously the transmission rate and the minimal distance.* From this perspective, good codes are isolated ones or lying close to the asymptotic bound. Below we briefly describe known results showing that “most” randomly chosen codes are not good. On the contrary, in the next section we show that in order to recognize good codes one must generate codes of low Kolmogorov complexity; that is, codes allowing short programs producing them.

This is exactly what has happened historically, when algebraic geometric codes, discovered by Goppa, were used by Tsfasman, Vladut, and Zink in order to ameliorate the Gilbert–Varshamov bound: cf. an early survey [16] and [18].

In this sense, the moral of this note is just opposite to the title of [4]: *Only codes about which we can think can be good.*

**2.3. Shannon's ensemble.** We sketch here some well known arguments and results (see e.g. [4], [1], [18] and references therein) showing that most (unstructured)  $q$ -ary codes lie lower than or only slightly above the curve

$$(2.2) \quad R = \frac{1}{2}(1 - H_q(\delta))$$

where  $H_q(\delta)$  for  $0 < \delta < 1$  is the  $q$ -ary entropy function

$$H_q(\delta) = \delta \log_q(q-1) - \delta \log_q \delta - (1-\delta) \log_q(1-\delta).$$

Notice that only the part of this curve for which  $0 \leq R \leq 1/2$  lies inside  $[0, 1]^2$ .

The Gilbert–Varshamov bound GV

$$(2.3) \quad R = 1 - H_q(\delta)$$

plays a similar role for linear codes: cf. Remarks below.

In order to make the statements above precise, one introduces Shannon's Random Code Ensemble  $RCE_n$  of  $q$ -ary codes of block length  $n$ . Each code in  $RCE_n$  is a set of pairwise different words in  $A_q^n$  chosen randomly and independently with uniform probability  $q^{-n}$ .

**Proposition 2.2.** ([4], sec. V)

- (i) For any  $\varepsilon > 0$ , the probability (in  $RCE_n$ ) that  $H_q(d/n) \geq \max(1 - 2R, 0) + \varepsilon$ , where  $R = k/n$ , is bounded by  $q^{-\varepsilon n(1+o(1))}$  as  $n \rightarrow \infty$ .
- (ii) Similarly, the probability that  $H_q(d/n) \geq 1 - R + \varepsilon$  is bounded by  $e^{-q^{\varepsilon n(1+o(1))}}$  as  $n \rightarrow \infty$ .

**Strategy of the proof.** One easily sees that the number of words at a (Hamming) distance  $\leq d$  from any fixed word in  $A_q^n$  is

$$(2.4) \quad \text{Vol}_q(n, d) = \sum_{j=0}^d \binom{n}{j} (q-1)^j.$$

As is well known (see [4]), one can estimate this quantity in terms of the  $q$ -ary entropy:

$$q^{(H_q(\delta) - o(1))n} \leq \text{Vol}_q(n, n\delta) \leq q^{H_q(\delta)n}.$$

Following [4], sec. V, denote by  $X^{(d)}$  the random variable on  $RCE_n$  whose value at a code is the number of unordered pairs of distinct code words at a distance  $\leq d$  from each other. Clearly, on codes of cardinality  $q^k$ , we have from (2.4)

$$E(X) = \binom{q^k}{2} \frac{\sum_{l=1}^d \binom{n}{l} (q-1)^l}{q^n - 1} = q^{n[H_q(d/n) - (1-2R)] + o(n)}.$$

One can similarly calculate  $E(X^2)$ , and then use Chebyshev's inequality to prove (i). The last statement is obtained along the same lines. For details, see [4].

**2.4. Remarks.** For unstructured code points  $(R, \delta)$  with  $1 - H_q(\delta) < 2R$ , the same reasoning shows that the average number of pairs at distance  $d$  is large.

In the case of linear codes, the relevant code points concentrate in an exponentially narrow neighborhood of the GV bound.

Since linear codes have considerably smaller Kolmogorov complexity than the general ones, this behavior is compatible with our discussion in Section 3 below.

**2.5. Spoiling operations as computable functions on codes.** The proof of existence of the asymptotic bound (essentially, the only known one) is based upon the existence of three types of rather banal combinatorial operations on (general, or linear) codes that produce from a given code several codes with worse parameters. The subsequent combinatorial characterization of isolated code points as points of finite multiplicity, and proof in [14] that any point with rational coordinates below the asymptotic bound is a code point, crucially uses these operations as well.

In each class, the result of the application of such an operation to a given code is by no means unique. Here, using the discussion in [15], sec. 1 (that reproduces several earlier sources), we will define three total recursive *spoiling maps*

$$S_i : Codes_q \rightarrow Codes_q, \quad i = 1, 2, 3,$$

that are compatible with the map  $C \mapsto [n(C), k(C), d(C)]_q$  and whose effect on code parameters is summarized below:

$$S_1 : [n, k, d]_q \mapsto [n + 1, k, d],$$

$$(2.5) \quad S_2 : [n, k, d]_q \mapsto [n - 1, k, d - 1], \quad (\text{if } n > 1, k > 0),$$

$$(2.6) \quad S_3 : [n, k, d]_q \mapsto [n - 1, k', d], \quad \text{where } k - 1 \leq k' < k, \quad (\text{if } n > 1, k > 1).$$

In order to make  $S_i$  unambiguous, we simply choose the code with minimal number with respect to some computable numbering from a finite set of codes that can be obtained in this way, and define it to be  $S_i(C)$ .

Another (minor) point is to decide what to do if the mild restrictions in brackets in (2.5), (2.6) do not hold for  $C$ . The simplest solution that we adopt is then to put  $S_i(C) = C$ .

**2.6. Block length and distance between isolated codes.** We will now show that knowing the distance of an isolated code point  $x$  from its closest neighboring code point, we can estimate from above the maximal block length of a code mapping to  $x$  and hence also the multiplicity of  $x$ . Distance in  $[0, 1]^2$  is defined here as  $\text{dist}((a, b), (c, d)) := \max(|a - c|, |b - d|)$ .

**Proposition 2.3.** *Let  $(R, \delta)$  be an isolated code point and denote by  $\rho$  its distance from the closest neighboring code point. Assume that  $(R, \delta)$  is the code point of some  $C \subset A^N$ . In this case we have*

$$N \leq \max\left(\frac{R - \rho}{\rho}, \frac{\delta - \rho}{\rho}\right).$$

**Proof.** If the code parameters of  $C$  are  $[N, K, D]_q$ , then the code parameters of  $S_1(C)$  are  $[N + 1, K, D]_q$  so that

$$\text{dist}(P_C, P_{S_1(C)}) = \max\left(\frac{[K]}{N + 1}, \frac{D}{N + 1}\right) \geq \rho.$$

This shows our result, because

$$R = \frac{[K]}{N}, \quad \delta = \frac{D}{N}.$$

### 3. Plotting the asymptotic bound with assistance of a complexity oracle

**3.1. A general setup.** Let  $X$  be an infinite constructive world, in the sense of [12]. This means that we have a set of *structural numberings* of  $X$ : computable bijections  $\mathbb{Z}^+ \rightarrow X$ , forming a principal homogeneous space over the group of total recursive permutations  $\mathbb{Z}^+ \rightarrow \mathbb{Z}^+$ .

Consider one such bijection  $\nu = \nu_X : \mathbb{Z}^+ \rightarrow X$ . It defines an order on  $X$ :  $x' \leq x$  iff  $\nu^{-1}(x') \leq \nu^{-1}(x)$ . Equivalently, we can imagine such a bijection as an order in which elements of  $X$  are generated:  $x$  is generated at the  $\nu(x)$ -th step.

Another important class of bijections that we have in mind consists of (uncomputable) *Kolmogorov orders* defined and discussed in [12]. Namely, let  $u : \mathbb{Z}^+ \rightarrow X$  be an optimal (in the sense of Kolmogorov and Schnorr) partial recursive enumeration. Then  $K_u(x) := \min\{k \mid u(k) = x\}$  is the (exponential) Kolmogorov complexity, and the Kolmogorov order of  $X$  is the order of growing complexity. If we denote the respective Kolmogorov order  $\mathbb{K}_u(x)$ , we have  $c_1 K_u(x) \leq \mathbb{K}_u(x) \leq c_2 K_u(x)$  for constants  $c_1, c_2 > 0$  depending only on  $u$ . Similarly, another choice of  $u$  produces another order differing from  $\mathbb{K}_u$  by a permutation of linear growth.

Let now  $X, Y$  be two infinite constructive worlds,  $\nu_X, \nu_Y$  respective structural bijections,  $u : \mathbb{Z}^+ \rightarrow X$ ,  $v : \mathbb{Z}^+ \rightarrow Y$  two optimal enumerations, and  $K_u, K_v$  the respective Kolmogorov complexities.

Consider a totally recursive function  $f : X \rightarrow Y$ .

**Proposition 3.1.** *For each  $y \in f(X)$ , there exists  $x \in X$  such that*

$$(3.1) \quad y = f(x), \quad K_u(x) \leq \text{const} \cdot \nu_Y^{-1}(y)$$

where the constant can be calculated in terms of  $u, v, \nu_X, \nu_Y$ .

**Proof.** Informally, this means that we can effectively generate all elements of the (enumerable) image  $f(X) \subset Y$  in their structural order, if an oracle generates for us all elements of  $X$  in the order of growing Kolmogorov complexity.

In fact, denote by  $F : X \rightarrow Y \times \mathbb{Z}^+$  the following map:

$$F(x) := (f(x), n(x)),$$

where

$$n(x) := \text{card} \{x' \mid \nu_X^{-1}(x') \leq \nu_X^{-1}(x), f(x') = f(x)\}.$$

In plain words,  $n(x)$  is the number of  $x$  in the set  $f^{-1}(f(x))$  with respect to the order induced by  $\nu_X$ .

Clearly,  $F$  is a (totally) recursive function. Hence the image  $E \subset Y \times \mathbb{Z}^+$  of  $F$  is an enumerable subset of  $Y \times \mathbb{Z}^+$ .

For each  $m \in \mathbb{Z}^+$ , put

$$X_m := \{x \in X \mid n(x) = m\} \subset X, \quad Y_m := f(X_m) \subset Y.$$

Then  $X_m$  (resp.  $Y_m$ ) is an enumerable subset of  $X$  (resp.  $Y$ ), and the restriction of  $f$  to  $X_m$  induces a bijection of these sets. Moreover,  $f(X_1) = f(X)$ , so that we can define the partially recursive function  $\varphi : Y \rightarrow X$ , with domain  $f(X)$ , which is  $f^{-1}$  on its domain.

Hence, for any  $x \in X_1$  such that  $f(x) = y$ , we will have

$$K_u(x) = K_u(\varphi(y)) \leq c_\varphi K_v(y) \leq c' \nu_Y^{-1}(y)$$

for appropriate constants  $c_\varphi, c'$ . Here and below we use some basic inequalities involving complexities, proved e.g. in [13], VI.9.

This completes the proof.

**3.2. Finite vs infinite multiplicity.** We keep notation of the previous subsections, in particular,  $f : X \rightarrow Y$  is a total recursive function. For any  $y \in Y$ , call

$$\text{mult}(y) := \text{card } f^{-1}(y)$$

the *multiplicity* of  $y$ . Proposition 3.1 shows that with assistance of a complexity oracle for  $X$  we can generate consecutively elements of  $Y$  of nonzero multiplicity. For applications to codes, we want to divide them into two basic subsets: elements of finite multiplicity  $Y_{fin}$  (they will correspond to isolated code points) and elements of infinite multiplicity  $Y_\infty$ . The latter will correspond to code points lying below or on the asymptotic bound.

From the definitions above, it is clear that

$$Y_\infty \subset \cdots \subset f(X_{m+1}) \subset f(X_m) \subset \cdots \subset f(X_1) = f(X),$$

and

$$Y_\infty = \bigcap_{m=1}^\infty f(X_m), \quad Y_{fin} = f(X) \setminus Y_\infty.$$

We have the following extension of Proposition 3.1.

**Proposition 3.2.** *For each  $y \in Y_\infty$  and each  $m \geq 1$ , there exists unique  $x_m \in X$  such that  $y = f(x_m)$ ,  $n(x_m) = m$ , and*

$$(3.2) \quad K_u(x_m) \leq \text{const} \cdot \nu_Y^{-1}(y)m \log(\nu_Y^{-1}(y)m)$$

where the constant does not depend on  $y, m$  and can be calculated in terms of  $u, v, \nu_X, \nu_Y$ .

**Proof.** Define the partial function

$$\Phi : Y \times \mathbb{Z}^+ \rightarrow X$$

with domain

$$D(\Phi) := \{ (y, m) \mid \text{mult}(y) \geq m \}$$

such that

$$\Phi(y, m) := \text{the } m\text{th element in the fiber } f^{-1}(y).$$

One easily sees that its graph is enumerable, hence  $\Phi$  is partial recursive. The element  $x_m$  in the statement of the proposition is just  $\Phi(y, m)$ . Therefore

$$K_u(x_m) = K_u(\Phi(y, m)) \leq \text{const} \cdot K((\nu_Y^{-1}(y), m)),$$

where  $K$  is some chosen Kolmogorov complexity of pairs. One can get various estimates of the Kolmogorov complexity  $K$  of the pair  $(\nu_Y(y), m)$  by choosing different structural numberings of the product of two constructive worlds  $Y \times \mathbb{Z}^+$ : cf. a more detailed discussion in Section 2.7–2.10 of [13]. Here we use the standard estimate symmetric in both arguments:

$$K((\nu_Y(y), m)) \leq \text{const} \cdot K(\nu_Y(y))K(m) \log(K(\nu_Y(y))K(m)).$$

This completes the proof, since complexity on  $\mathbb{Z}^+$  is majorized by identical function.

Now, an oracle mediated process of generating sets  $Y_\infty, Y_{fin}$ , can be described by the following inductive procedure. Choose a sequence of pairs of positive integers  $(N_m, m)$ ,  $m = 1, 2, 3, \dots$ ,  $N_{m+1} > N_m$ .

*Step 1.* Produce the list of all elements  $y \in f(X)$  such that  $\nu_Y^{-1}(y) \leq N_1$ . Denote this list  $A_1$ , and put  $B_1 = \emptyset$ .

If lists (subsets)  $A_m, B_m \subset f(X)$  are already constructed at the  $m$ -th step, go to

*Step  $m+1$ .* Produce the list of all elements  $y \in f(X)$  such that  $\nu_Y^{-1}(y) \leq N_{m+1}$ . Denote by  $A_{m+1}$  the subset of elements  $y$  in this list for which there exists  $x \in X$  with  $f(x) = y$ ,  $n(x) = m+1$ , and denote by



$B_{m+1}$  the subset of remaining elements. According to (3.2), we will have to ask the oracle to produce the list of  $x \in X$  with explicitly bounded complexity, in order to be sure that this  $x$  with  $n(x) = m + 1$  appears in this list, if it exists at all.

It is clear that  $A_m \cup B_m \subset A_{m+1} \cup B_{m+1}$ , and that the union of this sequence of sets is  $f(Y)$ . Moreover,  $B_m \subset B_{m+1}$  and  $\bigcup_{m=1}^{\infty} B_m = Y_{fin}$ . The passage from  $A_m$  to  $A_{m+1}$  generally involves both adding new elements  $y$  (with  $N_m < \nu_Y^{-1}(y) \leq N_{m+1}$ ) and forwarding some of the elements of  $A_m$  to  $B_{m+1}$  (whenever it turns out that in their fiber no new element of  $X$  appears).

**3.3. A structural numbering of  $q$ -ary codes.** We will now explain how the general procedure described above can be applied to codes. More precisely, we will describe the constructive world of  $q$ -ary codes  $X = Codes_q$ , the constructive world of rational points  $Y = [0, 1]^2 \cap \mathbb{Q}^2$ , and the total recursive map  $f : X \rightarrow Y$ ,  $C \mapsto cp(C)$ .

We fix  $q$  and the alphabet  $A$  of cardinality  $q$ . We fix also a total order on  $A$ , say, by identifying  $A$  with  $\{0, 1, \dots, q-1\}$ . We then order all words in  $A^n$  lexicographically.

Define now the following computable total order (or, equivalently, computable bijective enumeration  $\nu$ ) of the set of all nonempty codes  $Codes_q$  with  $k > 0$ :

- (a) If  $n_1 < n_2$ , all codes in  $A^{n_1}$  come before those in  $A^{n_2}$ .
- (b) If  $k_1 < k_2$ , all  $[n, k_1, d]_q$ -codes come before  $[n, k_2, d']_q$ -codes.
- (c) When  $(n, q^k)$  are fixed, order all words in  $A^n$  lexicographically, then consider the induced order on words in any code  $C \subset A^n$ , and finally encode  $C$  by the concatenation of all its elements put in lexicographic order. Such a word  $w(C) \in A^{nq^k}$  determines  $C$  uniquely. Finally, order all  $[n, k, d]_q$ -codes in the lexicographic order of  $w(C)$ .

Clearly,  $n(C), [k(C)] + 1$  and  $d(C)$  become total recursive functions  $Codes_q \rightarrow \mathbb{Z}_+$  (the condition  $k(C) > 0$  means that  $C$  contains at least two different words the distance between which is therefore positive).

Finally, the function “code point”

$$cp(C) := \left( \frac{[k]}{n}, \frac{d}{n} \right)$$

is a total recursive map  $Codes_q \rightarrow [0, 1]^2 \cap \mathbb{Q}^2$ .

**3.4. Plotting the asymptotic bound.** Now apply to the codes the general procedure discussed above. Fix an enumeration  $\nu_Y$  of rational points in the unit square in some natural way. To make the picture visually transparent, choose the sequence  $N_m$  in such a way that the set of points with  $\nu_Y^{-1}(y) \leq N_m$  contains the subset  $C_m$  of all points with

denominators of each coordinate dividing  $m!$ , and plot at the step  $m$  only the points of  $A_m, B_m$  contained in  $C_m$ .

Clearly, “pixels” in  $C_m$  form the vertices of the square lattice of radius  $1/m!$ . Call a subset of  $C_m$  *saturated* if it is a union of sets of the form  $S_{a,b} := \{(x, y) \mid x \leq a, y \leq b\}$ ,  $(a, b) \in C_m$ . To motivate this definition, recall that the subset of  $C_m$  lying under or on the asymptotic bound is saturated. Hence it must be contained in the maximal saturated subset  $D_m$  of  $A_m \cap C_m$ .

The upper boundary  $\Gamma_m$  of this subset (consisting of horizontal and vertical segments of the length  $1/m!$  that connect neighboring points) is our  $m$ -th approximation to the asymptotic bound.

Obviously,  $B_m$  is the subset of isolated codes constructed at the  $m$ -th step.

The status of any point that is above  $\Gamma_m$  but not contained in  $B_m$  will become clear only at a subsequent step.

#### 4. Partition functions for codes and phase transition effects

**4.1. Partition functions involving complexity.** Let  $X$  be an infinite constructive world. Following L. Levin ([7], [8]), we will consider functions

$$p : X \rightarrow \mathbb{R}_{>0} \cup \{\infty\}$$

that are *enumerable from below* in the sense that the set

$$\{(r, p(x)) \mid r < p(x)\} \subset \mathbb{Q} \times X$$

is enumerable.

Furthermore, Levin introduces the notion of a *quasinorm functional* on the set of enumerable from below functions and shows that for any choice of such a functional  $N$ , the set of functions  $p$  with  $N(p) < \infty$  admits a maximal one in the sense that it majorizes any other function after multiplication by an appropriate positive constant.

We quote here two representative examples showing the relation of this result to complexity:

**Proposition 4.1.** (i) Let

$$N(p) := \sup (r \cdot \text{card } \{x \mid p(x) \geq r\}).$$

For this quasinorm functional  $p(x) := K_u(x)^{-1}$  is a maximal function, where  $K_u$  is a Kolmogorov complexity on  $X$ .

(ii) Let

$$N(p) := \sum_{x \in X} p(x).$$

For this quasinorm functional,  $p(x) := KP_v(x)^{-1}$  is a maximal function, where  $KP_v$  is an (exponential) prefix-free complexity on  $X$  depending on the respective optimal “decompressor”  $v$ .

The initial definition of prefix-free complexity involved the choice of the world of binary words for  $X$ . However, Levin's result of Proposition 4.1(ii) gives a very natural independent characterization of this complexity in terms of *enumerable from below probability distributions* on  $X$  whose definition uses only the fact that  $X$  is a constructive world.

This construction shows that it is natural to consider at least two types of partition functions on a constructive world  $X$  that endow objects of low complexity with higher weight:  $Z(X, \beta) = \sum_{x \in X} K_u(x)^{-\beta}$  and  $ZP(X, \beta) = \sum_{x \in X} KP_v(x)^{-\beta}$  where  $\beta$  is the inverse temperature. Both absolutely converge for  $\beta > 1$  and diverge for  $\beta < 1$ . The difference is that the first one diverges at  $\beta = 1$ , whereas the second one converges for  $\beta = 1$  as well. Divergence is easily seen, if one replaces Kolmogorov complexity by Kolmogorov order (see [12], formula (3.10)), in which case the partition function becomes simply Riemann's  $\zeta(\beta)$ .

In the following we will use versions of  $Z(X, \beta)$ , in particular, because the usual Kolmogorov complexity was extended to the nonconstructive world of partial recursive functions (e.g. in [17] and in the first 1977 edition of [13]), and this allowed one to prove for it a number of useful estimates. Here is the simplest one.

**Lemma 4.2.** *Let  $Y$  be an infinite decidable subset of a constructive world  $X$ . Endowing  $Y$  with the induced structure of the constructive world, choose two exponential Kolmogorov complexities  $K_u(X, *)$ , resp.  $K_v(Y, *)$  of the objects in  $X$ , resp. in  $Y$ . Then the restriction of  $K_u(X, *)$  to  $Y$  is equivalent to  $K_v(Y, *)$  in the sense that one of these functions multiplied by a positive constant majorizes another one.*

**Proof.** The embedding  $i : Y \rightarrow X$  is a total recursive function. Define the function  $j : X \rightarrow Y$  as the identity on  $Y$ , and taking a constant value  $y_0 \in Y$  on the complement  $X \setminus Y$ . Since this complement is enumerable,  $j$  is total recursive as well. Hence  $K_u(i(y)) \leq c_1 K_v(y)$ ,  $K_v(j(x)) \leq c_2 K_u(x)$ .

**4.2. Phase transition.** Since the function  $\alpha_q(\delta)$  is continuous and strictly decreasing for  $\delta \in [1, 1 - q^{-1})$ , the limit points domain  $R \leq \alpha_q(\delta)$  can be equally well described by the inequality  $\delta \leq \beta_q(R)$  where  $\beta_q$  is the function inverse to  $\alpha_q$ .

Fix now an  $R \in \mathbb{Q} \cap (0, 1)$ . For  $\Delta \in \mathbb{Q} \cap (0, 1)$ , put

$$(4.1) \quad Z(R, \Delta; \beta) := \sum_{C: R(C)=R, \Delta \leq \delta(C) \leq 1} K_u(C)^{-\beta + \delta(C) - 1},$$

where  $K_u$  is an exponential Kolmogorov complexity on  $Codes_q$ .

**Theorem 4.3.** (i) *If  $\Delta > \beta_q(R)$ , then  $Z(R, \Delta; \beta)$  is a real analytic function of  $\beta$ .*

(ii) *If  $\Delta < \beta_q(R)$ , then  $Z(R, \Delta; \beta)$  is a real analytic function of  $\beta$  for  $\beta > \beta_q(R)$  such that its limit for  $\beta - \beta_q(R) \rightarrow +0$  does not exist.*

**Proof.** If  $\Delta > \beta_q(R)$ , then all codes in the summation domain of (4.1) are isolated ones, and there is only a finite number of them; hence  $Z(R, \Delta; \beta)$  is real analytic. Otherwise, this set of codes is an infinite decidable subset of  $Codes_q$ , and one can appeal to Lemma 4.2.

**4.2.1. Comments.** To embed the statement of Theorem 4.3 in a conventional environment of thermodynamics, one should have in mind the following analogies. The argument  $\beta$  in (4.1) is the inverse temperature, the transmission rate  $R$  is a version of density  $\rho$ , so that our asymptotic bound transported into the  $(T = \beta^{-1}, R)$ -plane as  $T = \beta_q(R)^{-1}$  becomes the phase transition boundary in the (temperature, density)-plane.

**4.3. Measures and the asymptotic bound.** We now show that the plotting procedure described in Section 3 can be reformulated in terms of measures on the space of codes defined by the partition functions described above.

The partition function  $Z(X, \beta) = \sum_{x \in X} K_u(C)^{-\beta}$  determines a one-parameter family of probability measures on the space  $X$  of codes, for  $\beta > 1$ , given by

$$\mathbb{P}_\beta(C) = \frac{K_u(C)^{-\beta}}{Z(\beta)}.$$

Similarly, one obtains probability measures associated to the partition functions  $ZP(X, \beta)$  and  $Z(R, \Delta; \beta)$ , with the latter defined on the space of codes with parameter  $R(C) = R$  and  $1 - \Delta \leq \delta(C) \leq 1$ .

Now consider again the oracle mediated process described in Section 3, generating the sets  $Y_\infty = V_q \cap U_q$  and  $Y_{fin} = V_q \setminus (V_q \cap U_q)$  of code points below and above the asymptotic bound, and the inductively constructed sets  $A_m$  and  $B_m$ .

**Proposition 4.4.** *The algorithm of Section 3 determines a sequence of probability measures associated to the sets  $A_m$  and  $B_m$  that converge to probability measures on the space of codes with parameters in  $Y_{fin}$  and  $Y_\infty$  and a sequence of measures  $\mathbb{P}_{m,\beta}$  converging to a measure supported on the asymptotic bound curve.*

**Proof.** We work with the partition function  $Z(X, \beta)$ . The argument for  $ZP(X, \beta)$  is analogous. On each of the sets  $B_m$  constructed by the oracle mediated algorithm of Section 3, one obtains an induced probability measure  $\mathbb{P}_{B_m,\beta}(C) = K_u(C)^{-\beta} Z(cp^{-1}(B_m), \beta)^{-1}$ , where

$$Z(cp^{-1}(B_m), \beta) = \sum_{C \in cp^{-1}(B_m)} K_u(C)^{-\beta}.$$

Since all the code points in  $Y_{fin}$  have finite multiplicity, and each  $B_m$  contains finitely many code points, the  $Z_{B_m}(\beta)$  are finite sums for all  $m \geq 1$ . Since the sets  $B_m \subset B_{m+1}$  are nested, in the limit  $m \rightarrow \infty$ ,

the probability measures  $\mathbb{P}_{B_m, \beta}(C)$  converge to the probability measure supported on  $cp^{-1}(Y_{fin})$  given by

$$\mathbb{P}_{fin, \beta}(C) = K_u(C)^{-\beta} Z(cp^{-1}(Y_{fin}), \beta)^{-1}$$

with  $Z(cp^{-1}(Y_{fin}), \beta) = \sum_{C \in cp^{-1}(Y_{fin})} K_u(C)^{-\beta}$ .

In the case of the sets  $A_m$ , one has  $A_m = (A_m \cap A_{m+1}) \cup (A_m \cap B_{m+1})$ , and one obtains the set  $Y_\infty = V_q \cap U_q$  of code points below the asymptotic bound as

$$Y_\infty = \bigcup_{m \geq 1} \left( \bigcap_{n \geq 0} A_{m+n} \right).$$

Consider the sequence of sets  $E_{M,N} = \bigcup_{m=1}^M \bigcap_{n=0}^N A_{m+n}$ . Then  $E_{M+1,N} \supset E_{M,N}$  and  $E_{M,N+1} \subset E_{M,N}$ . Correspondingly, one has sequences of probability measures

$$\mathbb{P}_{E_{M,N}}(C) = \frac{K_u(C)^{-\beta}}{Z(cp^{-1}(E_{M,N}), \beta)},$$

with  $Z(cp^{-1}(E_{M,N}), \beta) = \sum_{C \in cp^{-1}(E_{M,N})} K_u(C)^{-\beta}$ , that converge, as  $M, N \rightarrow \infty$  to the probability measure

$$\mathbb{P}_{\infty, \beta}(C) = K_u(C)^{-\beta} Z(cp^{-1}(Y_\infty), \beta)^{-1},$$

supported on codes in  $cp^{-1}(Y_\infty)$  with

$$Z(cp^{-1}(Y_\infty), \beta) = \sum_{C \in cp^{-1}(Y_\infty)} K_u(C)^{-\beta}.$$

Consider then the sets  $C_m$  and  $D_m \subset A_m \cap C_m$  constructed as in Section 3, and the upper boundary  $\Gamma_m$  of the set  $D_m$  approximating the asymptotic bound. Denote by  $F_m \subset D_m$  the region bounded between  $\Gamma_m$  and  $D_m \cap \Gamma_{m-1}$ . Then the probability measures

$$\mathbb{P}_{F_m}(C) = \frac{K_u(C)^{-\beta}}{Z(cp^{-1}(F_m), \beta)}$$

with  $Z(cp^{-1}(F_m), \beta) = \sum_{C \in cp^{-1}(F_m)} K_u(C)^{-\beta}$  converge to a probability measure  $\mathbb{P}_\Gamma(C) = K_u(C)^{-\beta} Z(\Gamma, \beta)^{-1}$ , supported on the set of codes whose code points fall on the asymptotic bound curve  $\Gamma = \{(R, \delta) \mid R = \alpha_q(\delta)\}$  itself, with  $Z(\Gamma, \beta) = \sum_{C \in cp^{-1}(\Gamma)} K_u(C)^{-\beta}$ .

When using the partition function  $Z(R, \Delta; \beta)$  of (4.1), one has an analogous statement, with code points restricted to the domain  $R(C) = R$  and  $1 - \Delta \leq \delta(C) \leq 1$ , except for the last statement about a measure supported at the asymptotic bound, because of the presence of a phase transition for  $Z(R, \Delta; \beta)$  precisely along that curve.

## 5. From classical to quantum systems

**5.1. Computable functions as classical observables.** In subsection 4.3 we have described a statistical mechanical system on the space of codes, where observables are computable functions on the space  $X$  of  $q$ -ary codes and the expectation values of observables are obtained by integrating these functions against the probability measure defined by the complexities,

$$\langle f \rangle_\beta = \int f(C) d\mathbb{P}_\beta(C) = \frac{1}{Z(X, \beta)} \sum_{C \in X} f(C) K_u(C)^{-\beta},$$

or similarly with the measures defined by  $ZP(X, \beta)$  or  $Z(R, \Delta, \beta)$ .

In this section we describe a quantized version of this statistical system and explain the role of the asymptotic bound curve  $R = \alpha_q(\beta)$  in this setting.

The quantization of the system is achieved by considering code words as the possible independent degrees of freedom in an unstructured code, and quantizing them as independent harmonic oscillators, with energy levels that depend on the rate and the complexity of the code.

We then show that, while code points that lie below the asymptotic bound give rise in this way to a bosonic field theory with infinitely many degrees of freedom, the code points above the asymptotic bound produce quantum mechanical systems with finitely many degrees of freedom.

The partition function of the resulting quantum statistical mechanical system is different from the one we computed in Section 3 for the classical system, but it is easily derived from it and displays similar phase transition phenomena. We remind that the quantum partition function is essentially  $\text{Tr}(e^{-\beta H})$  where  $H$  is the relevant Hamiltonian operator.

We also show that the recursive algorithm of Section 2 provides a good approximation by systems with finitely many degrees of freedom for the quantum system associated to the set  $Y_\infty$  of code parameters.

## 5.2. Quantum statistical mechanical system of a single code.

To make a single unstructured code  $C$  into a quantum system, we regard the code words as the possible independent degrees of freedom and we associate to each of them a creation and annihilation operator. This means that we consider for each code word  $x \in C$  an isometry  $T_x$ , with  $T_x^* T_x = 1$  and such that the  $T_x T_x^*$  are mutually orthogonal projectors. This means that we associate to a given code the Toeplitz algebra  $\mathcal{T}_C$  on its set of code words. This is the same kind of code algebra as we considered in our previous work [15].

The algebra  $\mathcal{T}_C$  is naturally represented on the corresponding Fock space, namely the Hilbert space  $\mathcal{H}_C$  spanned by the orthonormal basis

$\epsilon_w$  with  $w = x_1 \dots x_N$  ranging over the set of finite sequences (of arbitrary length  $N$ ) of the code words  $x \in C$ . In this representation, the operator  $T_x$  acts by appending  $x$  as a prefix to a given string of code words,  $T_x \epsilon_w = \epsilon_{xw}$ .

The dynamics of this quantum system is determined by a Hamiltonian operator  $H$  on the Fock space, via the time evolution

$$T \mapsto q^{itH} T q^{-itH}.$$

We can then assign energy levels that depend on the complexity of the code in the following way.

**Lemma 5.1.** *The time evolution  $\sigma : \mathbb{R} \rightarrow \text{Aut}(\mathcal{T}_C)$  given by  $\sigma_t(T_x) = K_u(C)^{it} T_x$  is generated by the Hamiltonian  $H\epsilon_w = \ell(w) \log_q K_u(C) \epsilon_w$ , with  $\ell(w)$  the length of the word  $w$ , and has partition function*

$$(5.1) \quad Z(C, \sigma, \beta) = \frac{1}{1 - q^{nR} K_u(C)^{-\beta}},$$

which is a real analytic function in the domain  $\beta > nR / \log_q K_u(C)$ .

**Proof.** The Hamiltonian implementing the time evolution  $\sigma_t(T_x) = K_u(C)^{it} T_x$  on the Fock space  $\mathcal{H}_C$  is the operator  $H$  on  $\mathcal{H}_C$  satisfying

$$\sigma_t(A) = q^{itH} A q^{-itH}, \quad \forall A \in \mathcal{T}_C.$$

This is given by the operator  $H\epsilon_w = m \log_q K_u(C) \epsilon_w$  for all words  $w = x_1 \dots x_m$  of length  $\ell(w) = m$ . We then find

$$\begin{aligned} Z(C, \sigma, \beta) &= \text{Tr}(q^{-\beta H}) = \sum_m (\text{card } W_m) q^{-\beta m \log_q K_u(C)} \\ &= \sum_m q^{m(nR - \beta \log_q K_u(C))}, \end{aligned}$$

where the cardinality of the set  $W_m$  of words of length  $m$  is  $q^{mk}$ , since  $\text{card } C = q^k = q^{nR}$ , with  $n(C) = n$  the length of the code and  $R(C) = R$  the rate. This series converges to (5.1) for  $\beta > nR / \log_q K_u(C)$ .

To compare the behavior of this partition function to the  $Z(R, \Delta; \beta)$  considered in Theorem 4.3, it is convenient to change variable in (5.1) by  $\beta \mapsto n(C)(\beta - \delta(C) + 1)$ . Then, using the Singleton bound on codes, we obtain the following.

**Corollary 5.2.** *The function  $Z(C, \sigma, \alpha)$ , with  $\alpha = n(\beta - \delta + 1)$ , is real holomorphic for all  $\beta > 0$  with*

$$Z(C, \sigma, n(\beta - \delta + 1)) \leq \frac{1}{1 - K_u(C)^{-\beta}}.$$

**Proof.** The partition function  $Z(C, \sigma, \alpha)$  is given by the sum

$$\begin{aligned} \sum_m q^{mn(R-(\beta-\delta+1)\log_q K_u(C))} &\leq \sum_m q^{mn(R+\delta-1)} K_u(C)^{-\beta m} \\ &\leq \sum_m K_u(C)^{-\beta m}, \end{aligned}$$

where the first estimate uses  $K_u(C)^{\delta-1} \leq q^{\delta-1}$  and the second estimate uses the singleton bound for codes  $k \leq n-d-1$ , which gives  $R+\delta-1 \leq 0$ .

**5.3. Quantum statistical mechanical system at a fixed code point.** We can now consider quantum statistical mechanical systems involving several codes. Again, the main idea is that different codes with their degrees of freedom given by their code words are considered as independent uncoupled systems, which means that the algebra of observables describing the set  $X_{(R,\delta)}$  of  $q$ -ary codes with fixed code parameters  $(R, \delta)$  becomes the tensor product of the Toeplitz algebras of the individual codes,

$$(5.2) \quad \mathcal{T}_{(R,\delta)} = \otimes_{C \in X_{(R,\delta)}} \mathcal{T}_C$$

acting on the tensor product of the Fock spaces  $\mathcal{H}_{(R,\delta)} = \otimes_{C \in X_{(R,\delta)}} \mathcal{H}_C$  and with the product time evolution  $\sigma_t^{(R,\delta)} = \otimes_C \sigma_t^C$ , with  $\sigma_t^C(T_x) = K_u(C)^{it} T_x$ . The partition function becomes, correspondingly, the product of the partition function for the independent systems. In particular, for  $\alpha = \alpha(n, \beta, \delta) = n(\beta - \delta + 1)$  a variable inverse temperature as in [15], we find that

$$Z(X_{(R,\delta)}, \sigma, \alpha) = \prod_{C \in X_{(R,\delta)}} Z(C, \sigma, n(\beta - \delta + 1))$$

is a finite product for  $(R, \delta) \in Y_{fin}$  and an infinite product for  $(R, \delta) \in Y_\infty$ , whose convergence is controlled by the convergence of the infinite product

$$\prod_{C \in X_{(R,\delta)}} (1 - K_u(C)^{-\beta})^{-1}.$$

This in turn converges whenever the series

$$Z(X_{R,\delta}, \beta) = \sum_{C \in X_{(R,\delta)}} K_u(C)^{-\beta}$$

converges, which is the classical partition function for a fixed code point.

This argument shows the role of the asymptotic bound from the point of view of these quantized systems. In fact, recall that an infinite tensor product of Toeplitz algebras is a standard way to describe the second quantization of a bosonic field theory; see for instance [6] and also Section 2 of [2]. However, a finite tensor product is a purely quantum mechanical system, with only finitely many degrees of freedom. Thus,



the asymptotic bound separates the region  $Y_{fin}$  in the space of code parameters where the quantum statistical system  $(\mathcal{T}_{(R,\delta)}, \sigma_t)$  is purely quantum mechanical (first quantization) from the region  $Y_\infty$  where it is a second quantization of a bosonic field.

**5.4. Oracle assisted QSM system construction.** It is usually interesting in quantum statistical mechanics to construct explicit approximations to systems with infinitely many degrees of freedom by systems involving finitely many ones. The oracle mediated construction described in Section 3 provides us with this kind of procedure. Consider the sets  $A_m$  and  $B_m$  described in Section 3. We can then consider the algebras

$$\mathcal{A}_m = \otimes_{C \in cp^{-1}(A_m)} \mathcal{T}_C, \quad \text{and} \quad \mathcal{B}_m = \otimes_{C \in cp^{-1}(B_m)} \mathcal{T}_C,$$

acting on the tensor product of the Fock spaces, and endowed with the tensor product time evolution as above. Moreover, by denoting by  $F_m$  (as in the previous section) the region between the curves  $\Gamma_m$  and  $D_m \cap \Gamma_{m-1}$ , one can consider the QSM system associated to the codes with code points in  $F_m$ ,

$$\mathcal{F}_m = \otimes_{C \in cp^{-1}(F_m)} \mathcal{T}_C, \quad \sigma_t = \otimes \sigma_t^C.$$

These give good approximations, by systems involving only finitely many degrees of freedom, to the bosonic field theory associated to the set  $Y_\infty$  of code parameters and to the asymptotic bound  $\Gamma$ .

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